Light and Shading

EECS 442 – Prof. David Fouhey

Winter 2019, University of Michigan

http://web.eecs.umich.edu/~fouhey/teaching/EECS442_W19/

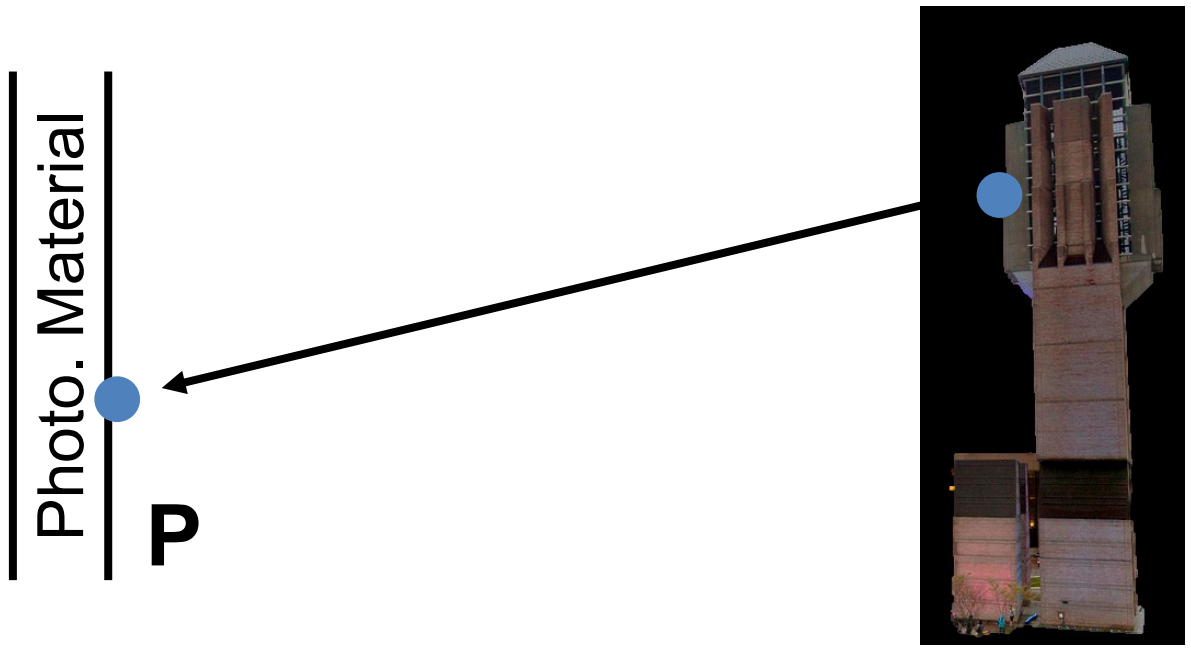
Administrivia

- I sent out requests for waitlist additions. I will continue to add as spots free up but chances are diminishing.
- ***PLEASE SIGN UP ON PIAZZA.*** There are no secrets on canvas.
- HW1 out. **Any general questions (not about content)?**
- Discussion on Wednesday: image processing / numpy. Materials out on piazza.

Recap: Projection

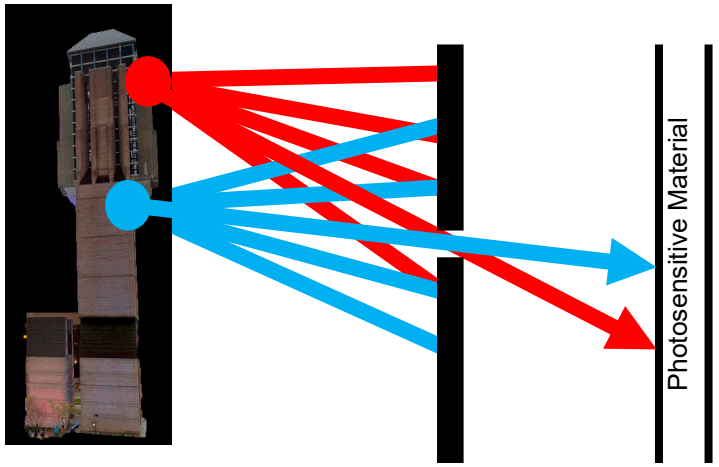
Image → $\boldsymbol{P} = \boldsymbol{K}[\boldsymbol{R}, \boldsymbol{t}]\boldsymbol{X}$ ← *World*

Intrinsic *Extrinsic*



Recap: Lenses

Pinhole Model



Mathematically correct
Not quite correct in practice
Reasonable approximation

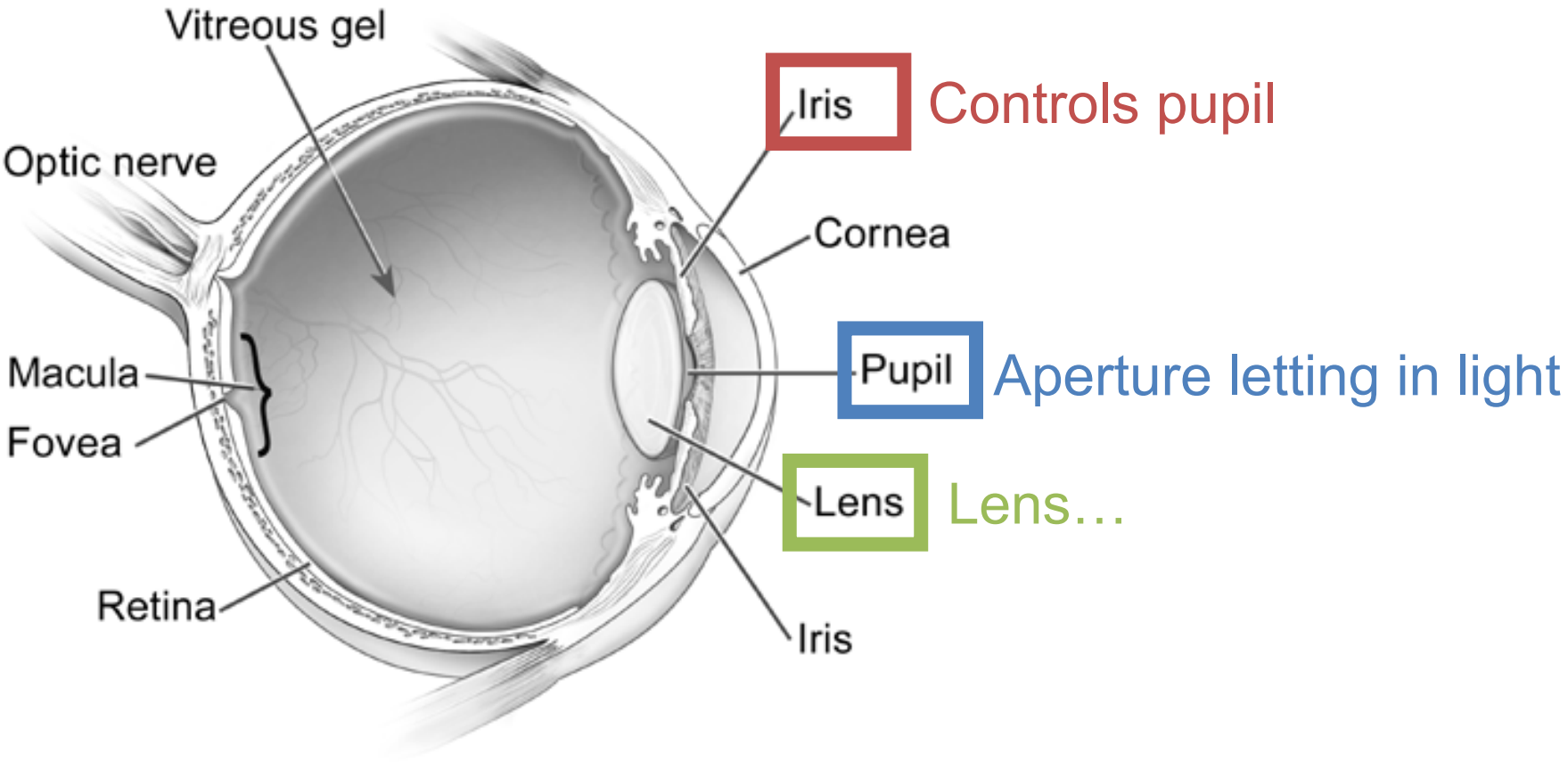
Reality: Lenses

Necessary in practice
Introduce complications
Complications fixable

Today

- A little bit about light and how you represent it
- A little bit about lighting and how it works

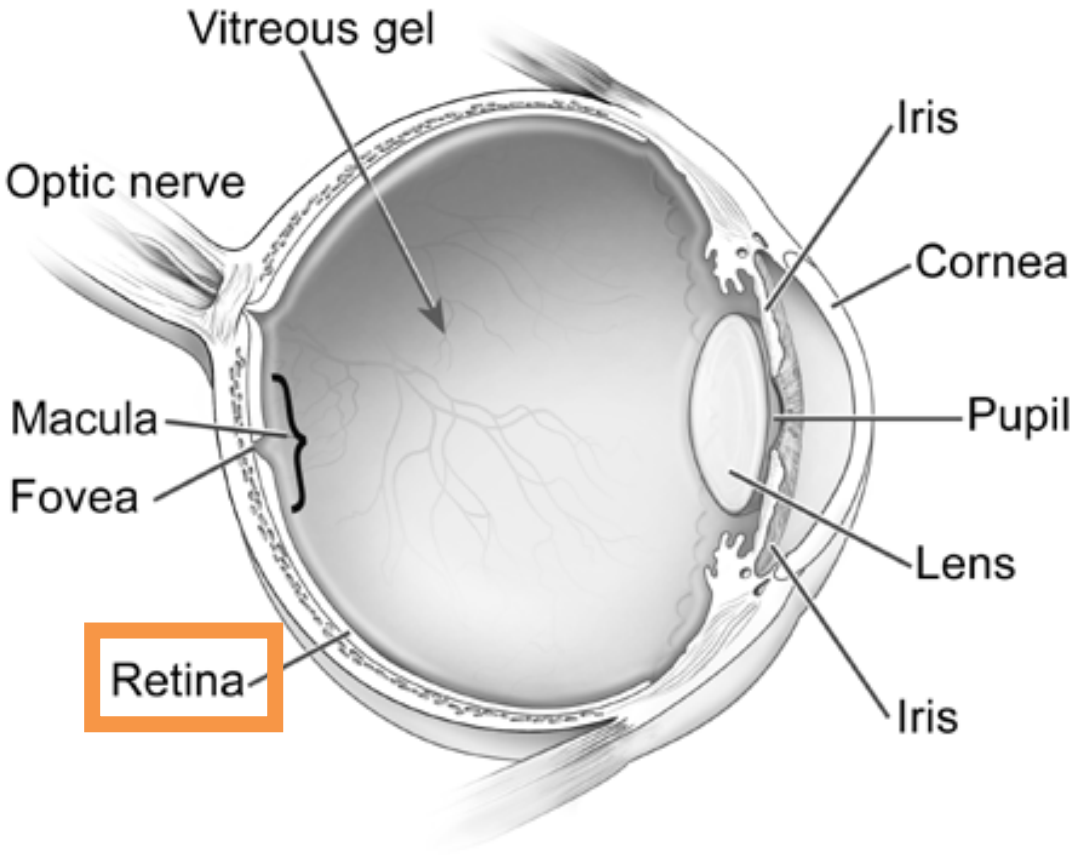
Your Very Own Camera



Where's the film/CCD?

Slide Credit: NIH

Your Very Own Camera

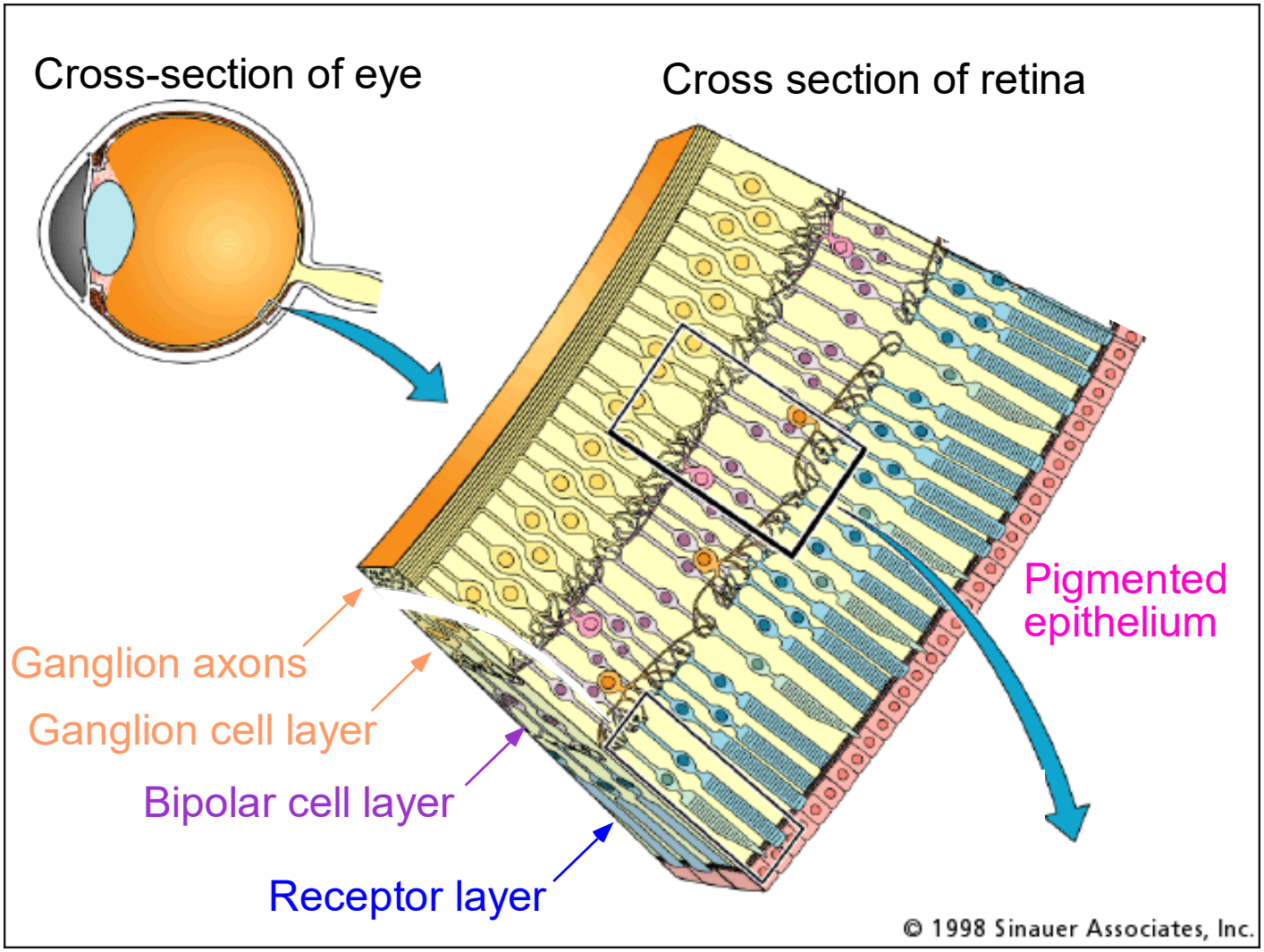


Where's the film/CCD?

Slide Credit: NIH

Demo Time

What is Retina/Film Made Of?



Slide Credit: J. Hays

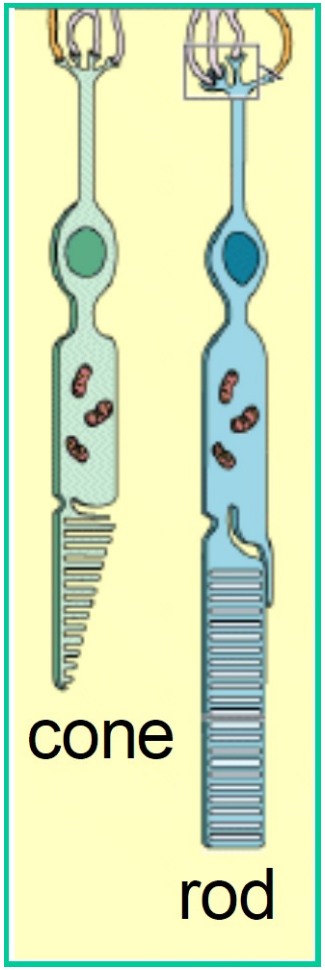
Two Type of Photo Receptors

Cones

cone-shaped
less sensitive
operate in high light
color vision

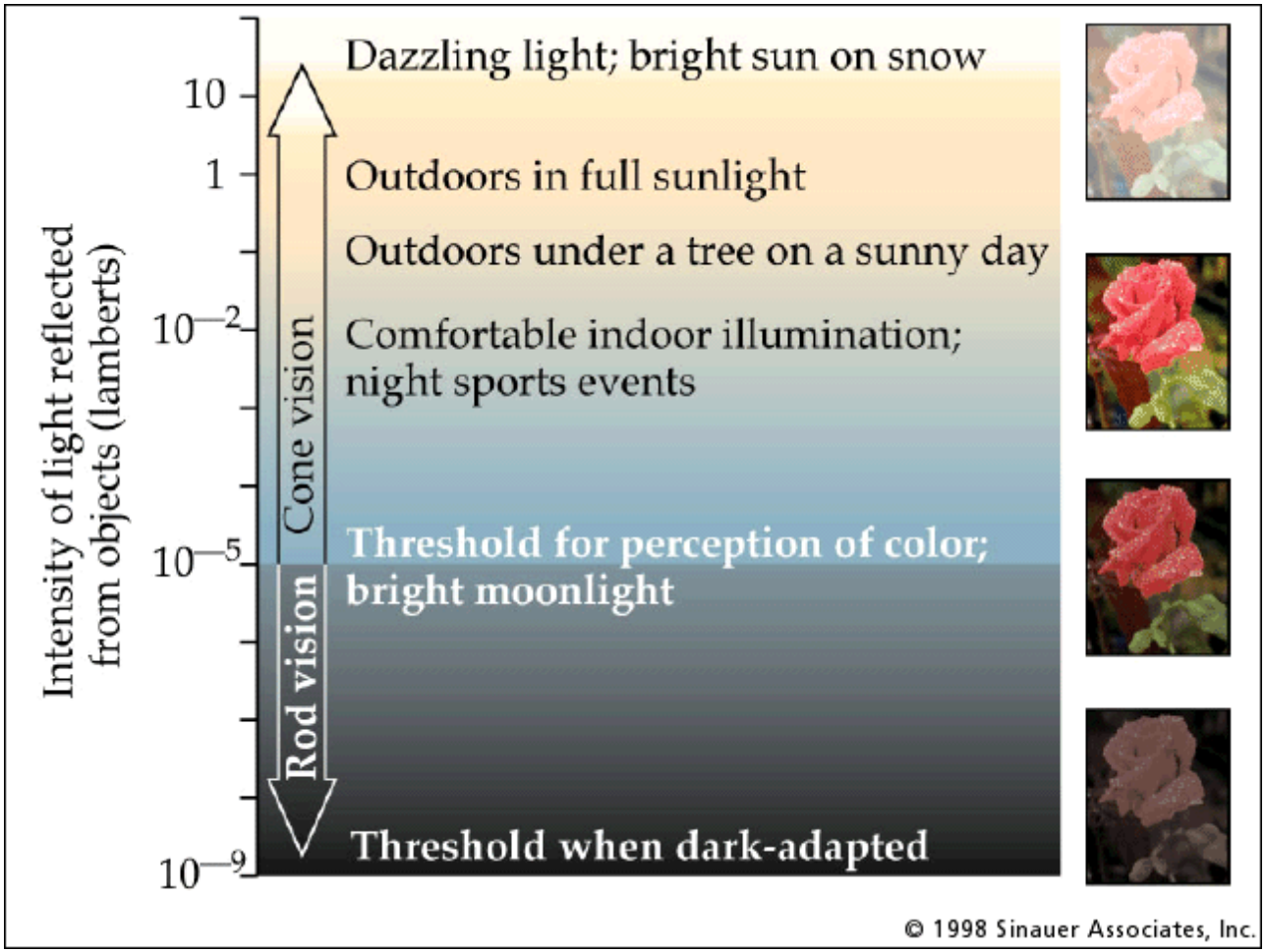
Rods

rod-shaped
highly sensitive
operate at night
gray-scale vision



Slide Credit: J. Hays

Rod / Cone Sensitivity



Slide Credit: J. Hays

Rod/Cone Distribution

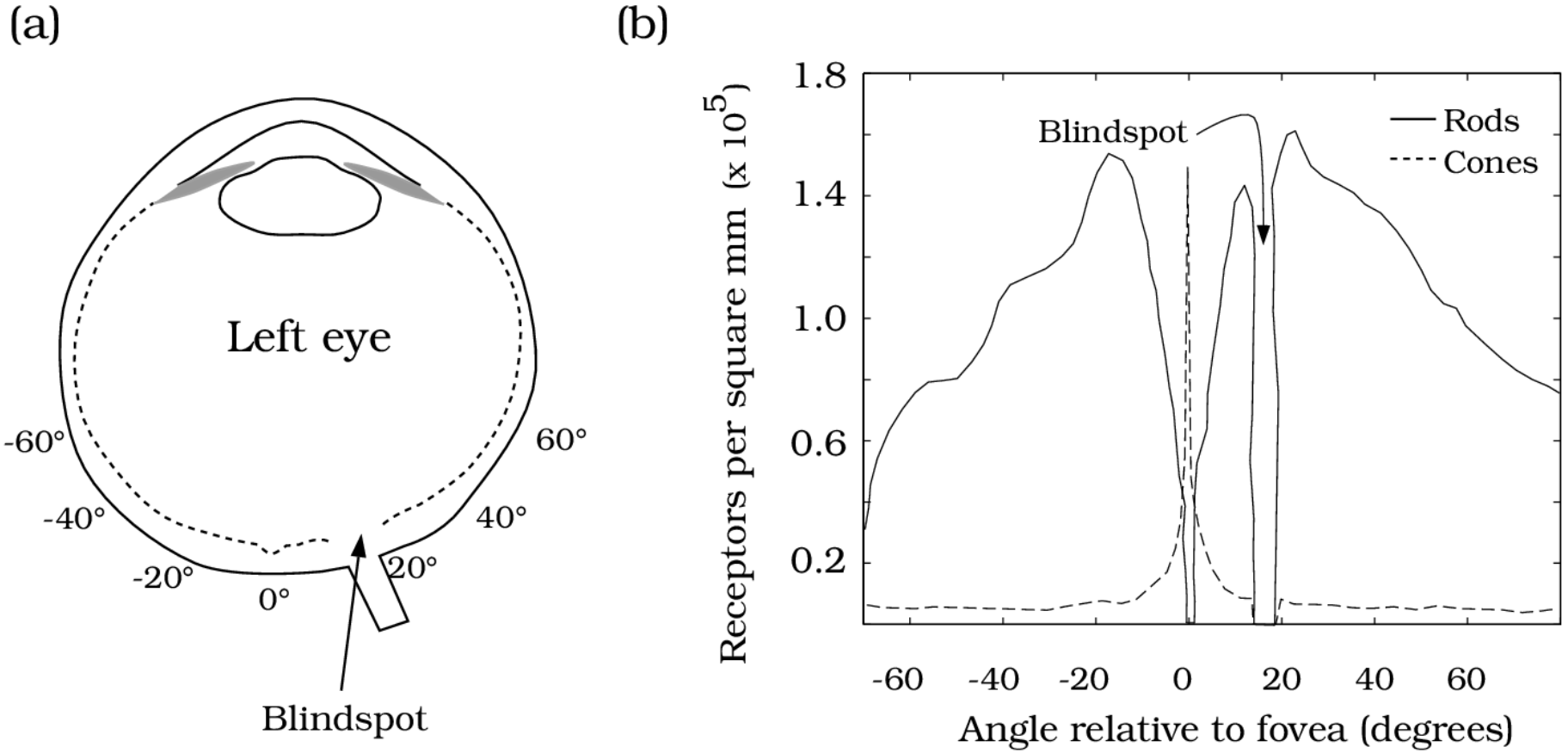
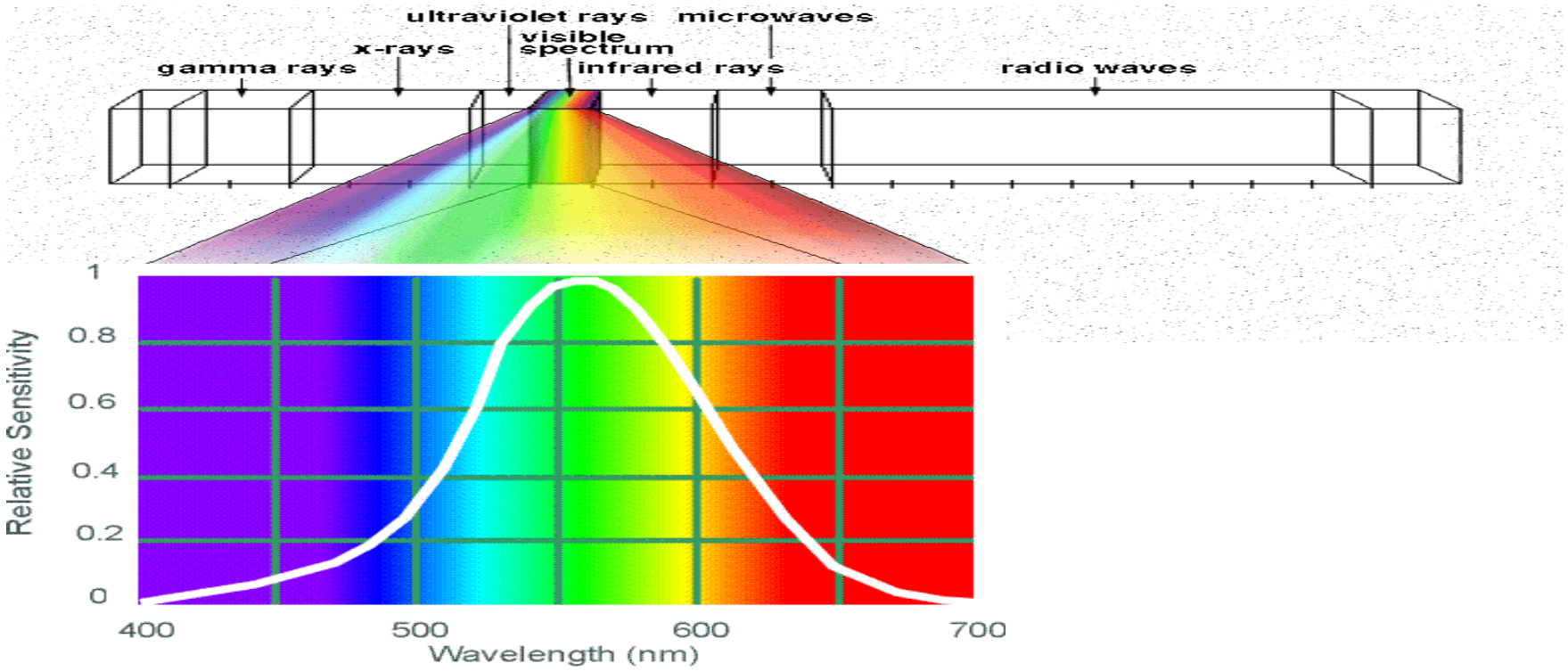


Diagram Credit: B. A. Wandell, *Foundations of Vision*

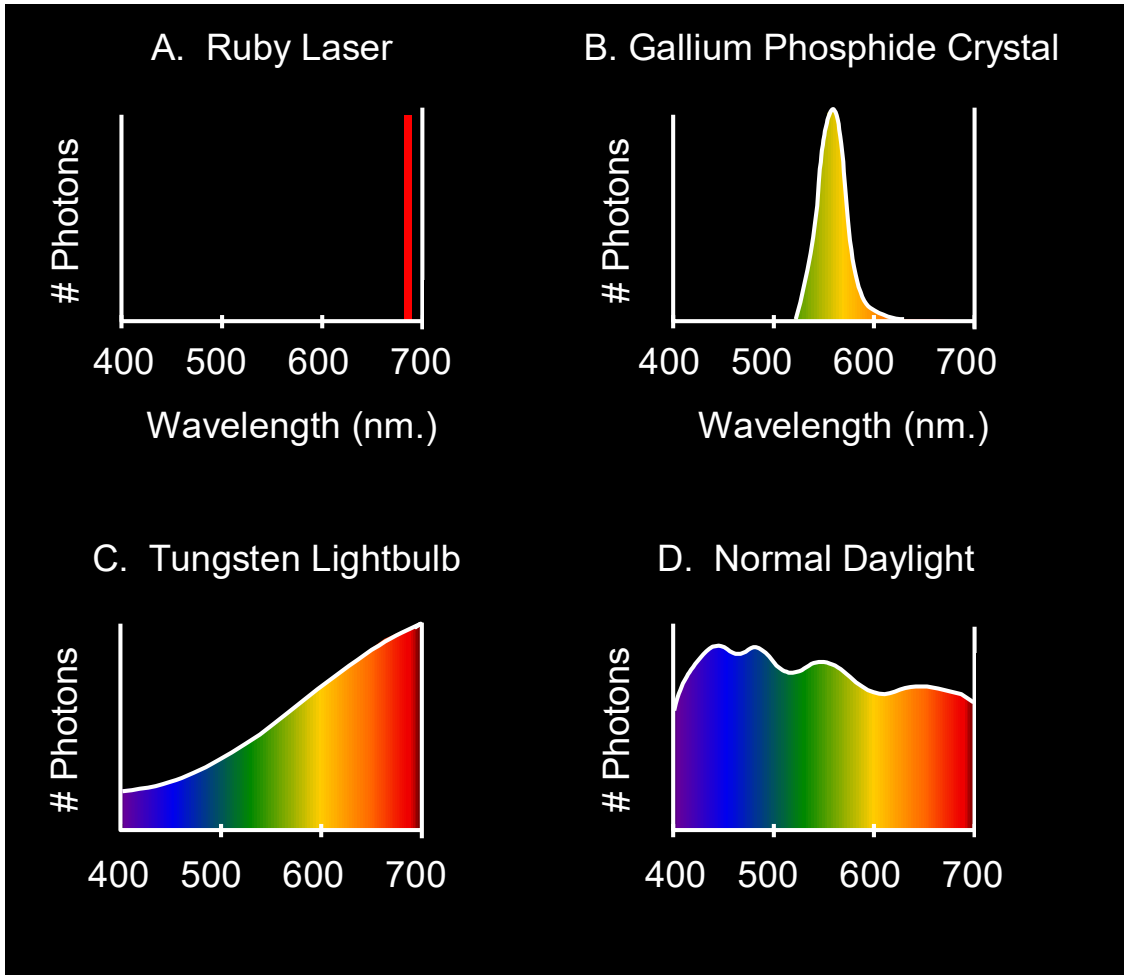
Electromagnetic Spectrum



Why do we see light in these wavelengths?

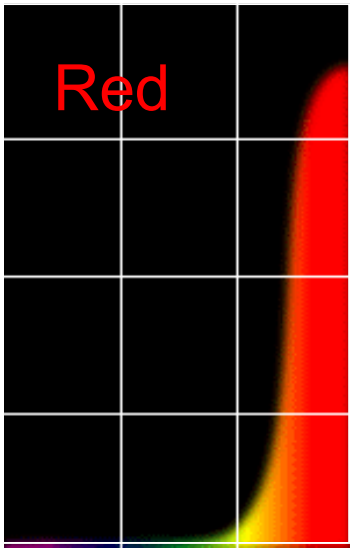
Slide Credit: J. Hays

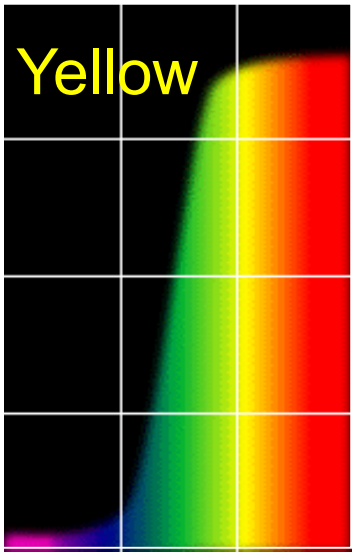
The Physics of Light

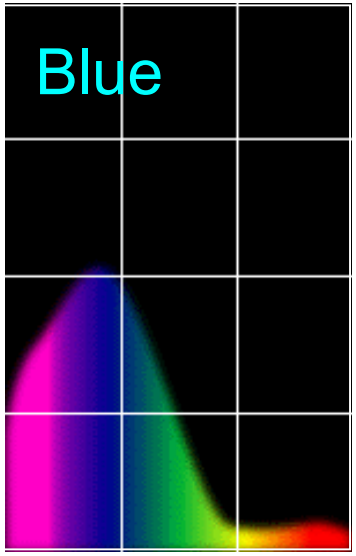


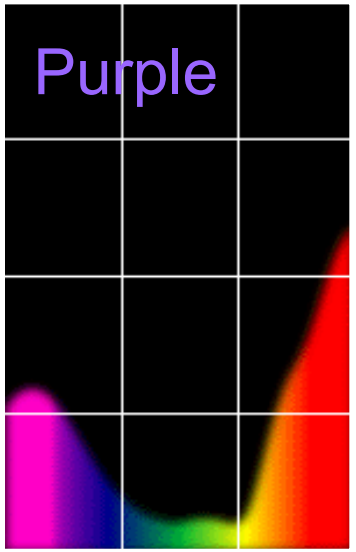
The Physics of Light



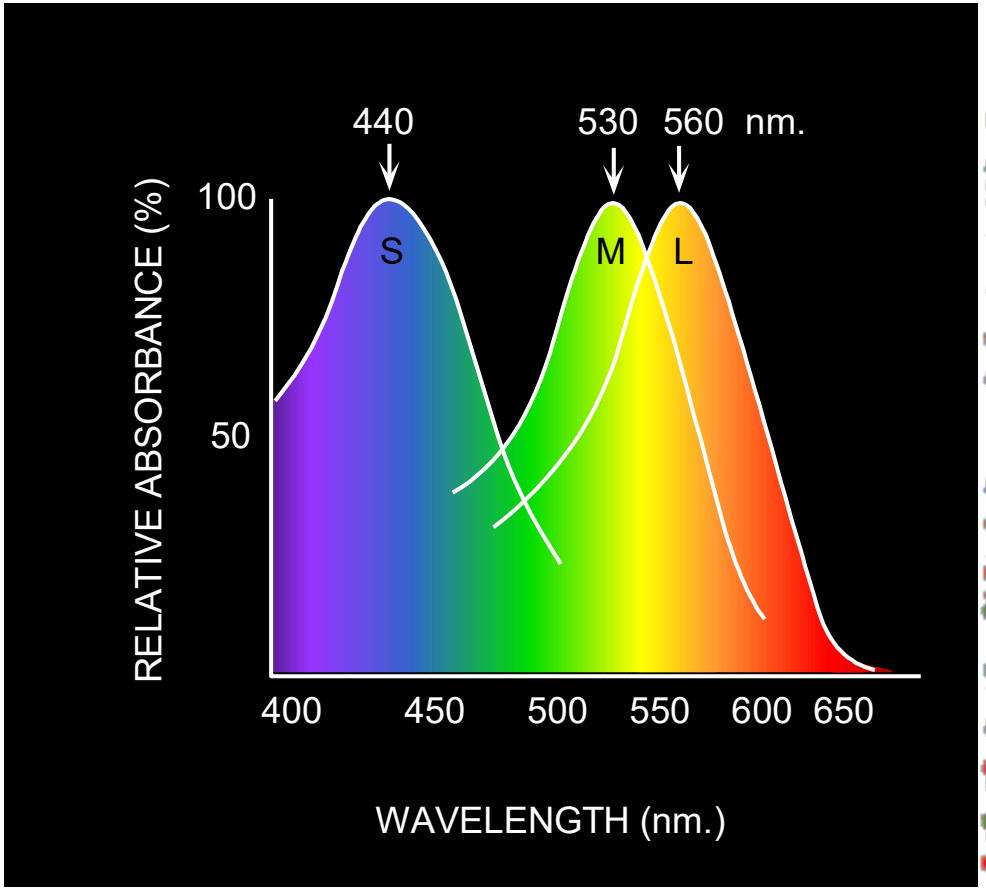



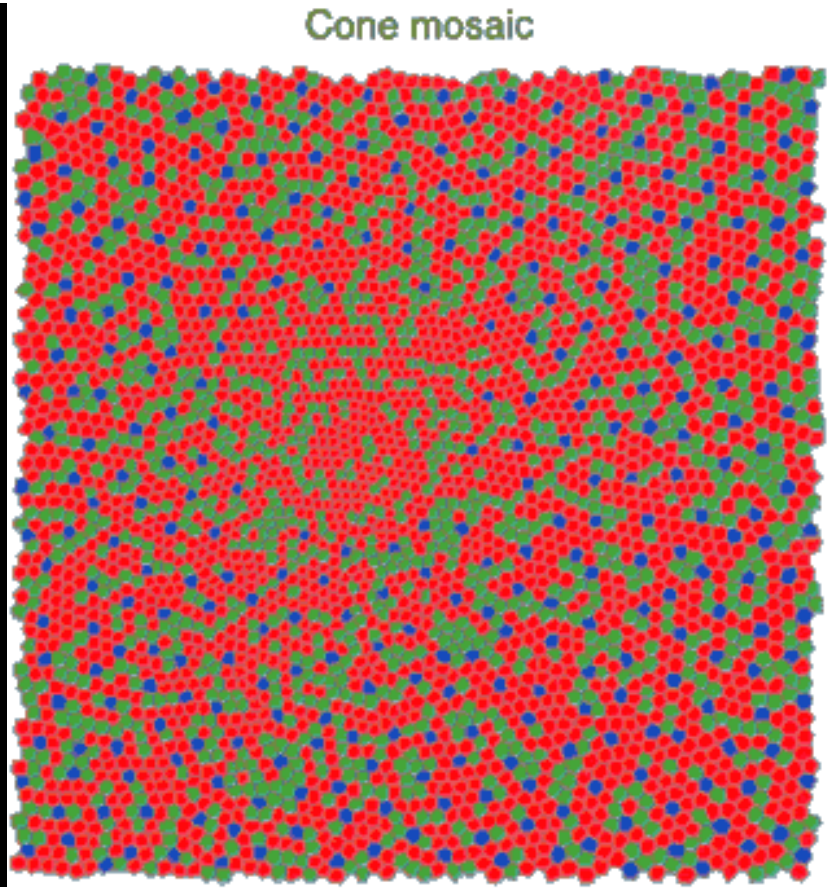






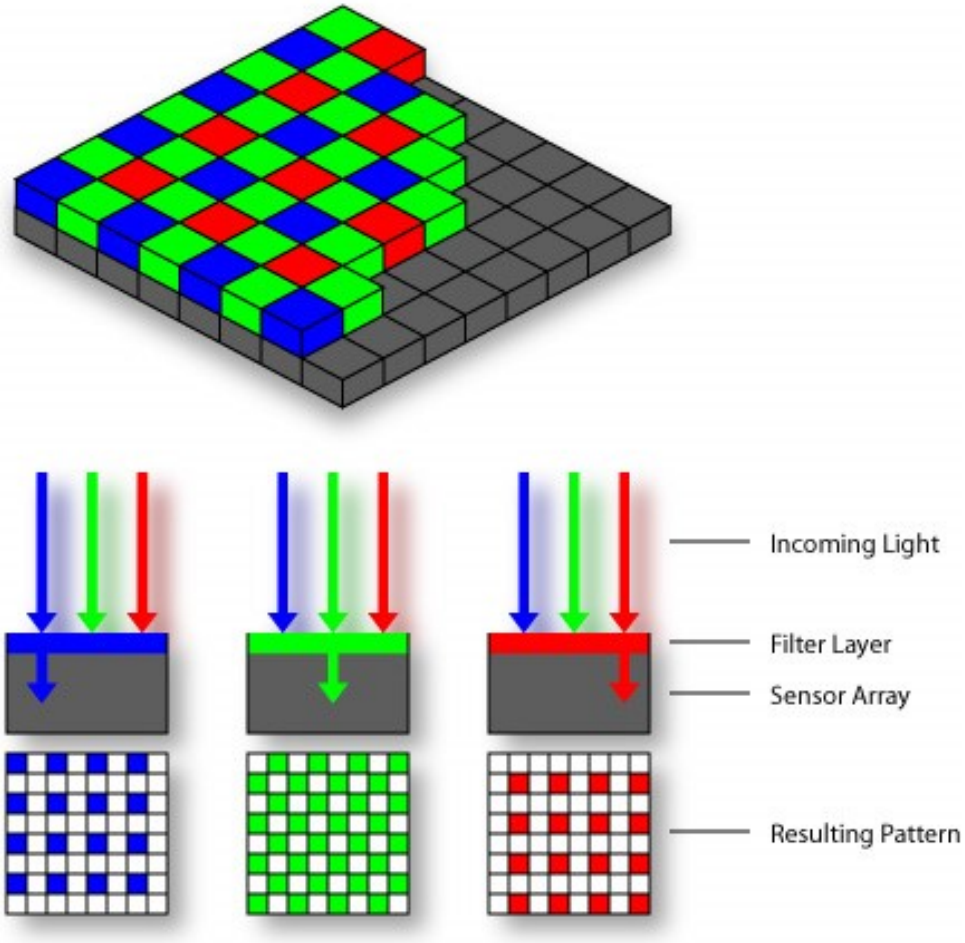
The Physics of Light





How Do We Get Light?

Artificial Cones

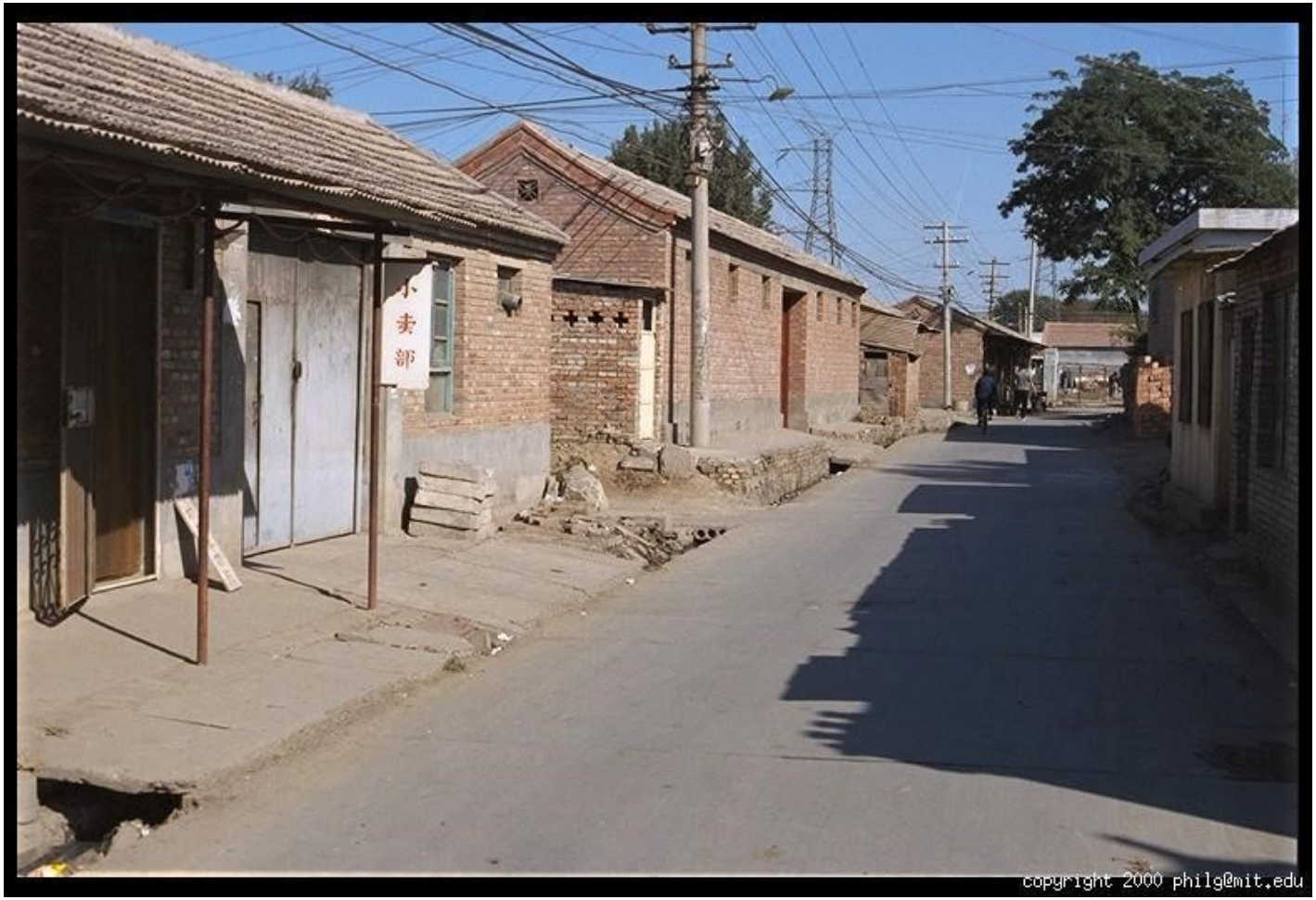


Estimate RGB at ‘G’ cells from neighboring values

Slide Credit: S. Seitz

Slide by Steve Seitz

Color Image



Slide Credit: J. Hays

Color Image

Combined

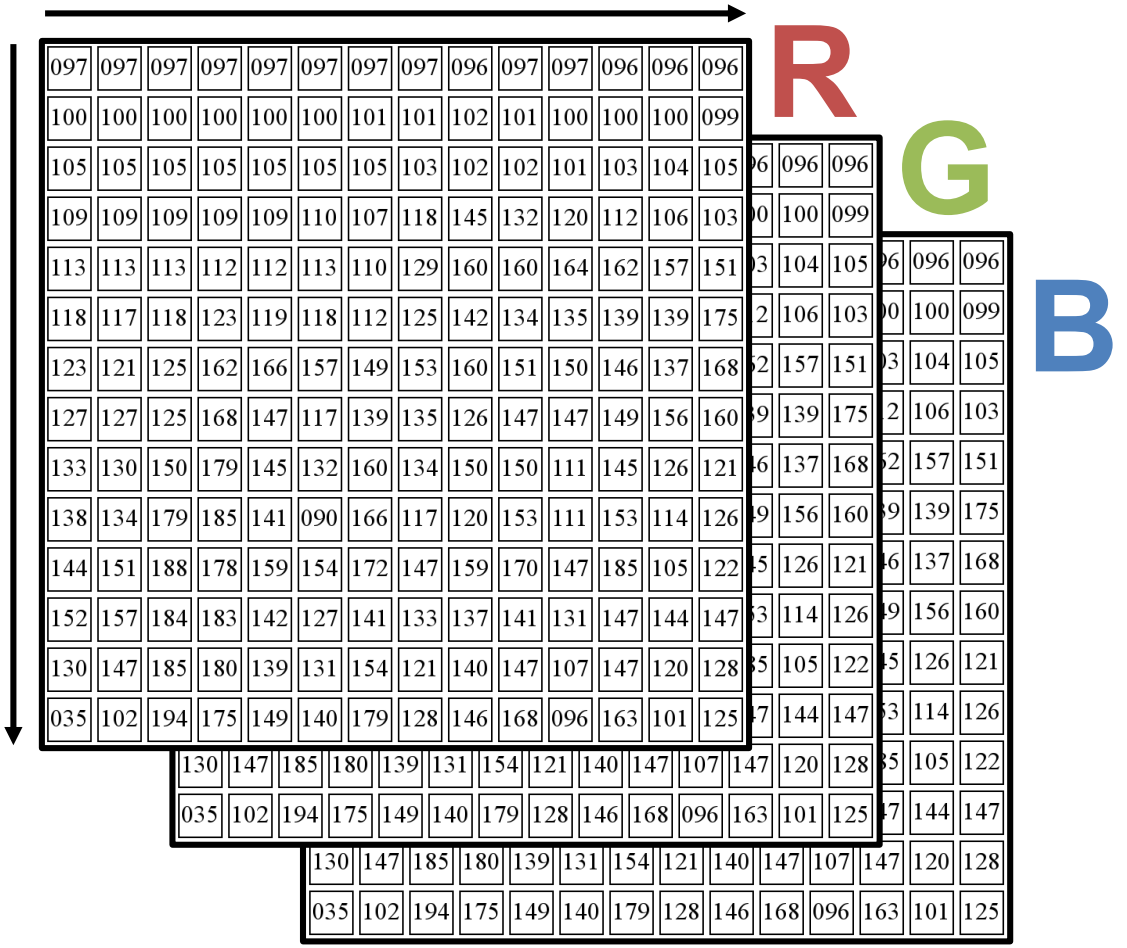
Red

Green

Blue

Slide Credit: J. Hays

Images in Python



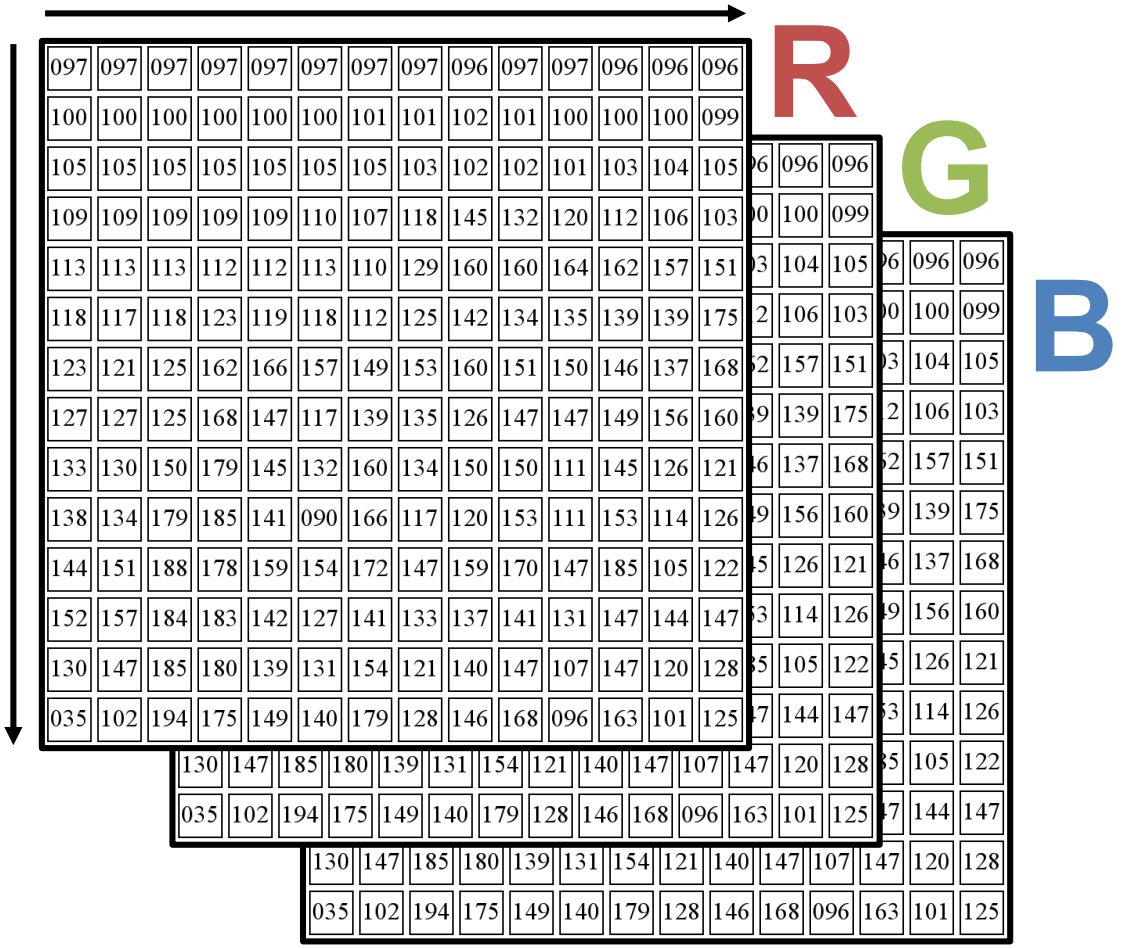
Images in Python

Images are matrix / tensor `im`

`im[0,0,0]`
top, left, red

`im[y,x,c]`
row y, column x, channel c

`im[H-1,w-1,2]`
bottom right blue

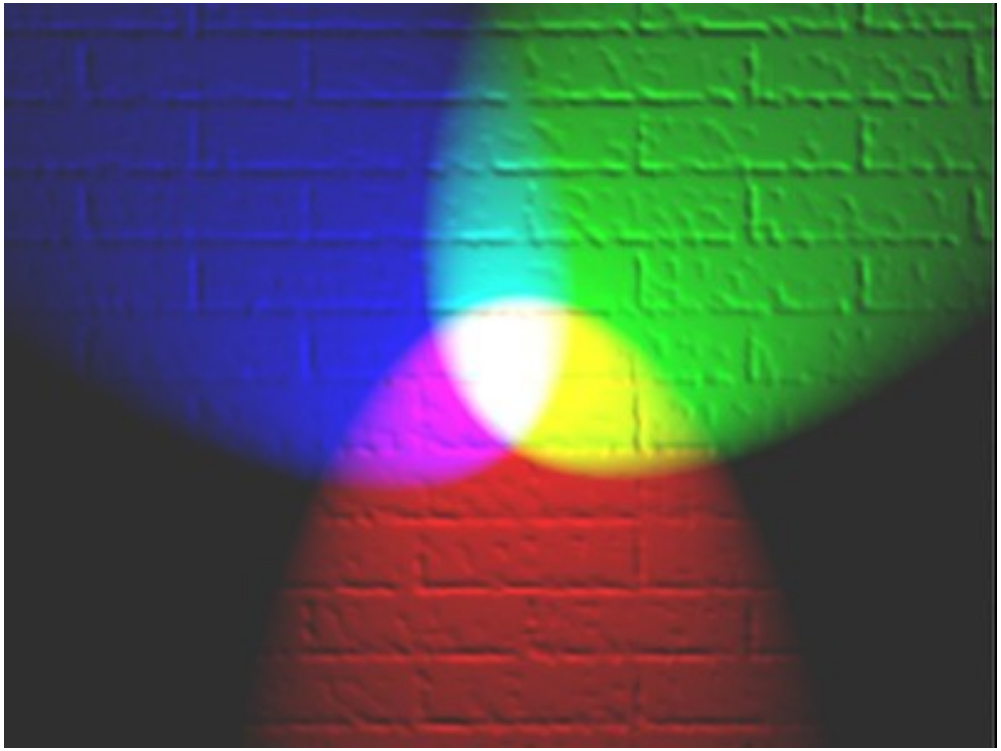


Slide inspired by James Hays

5 Things To Always Remember

1. Origin is top left
2. Rows are first index (**what's the fastest direction for accessing?)**
3. Usually referred to as Height x Width
4. Typically stored as uint8 [0,255]
5. for y in range(H): for x in range(W): will run 1 million times for a 1000x1000 image. *A 4GHz processor can do only 4K clock cycles per pixel per second.*

Representing Colored Light



Discussion time: how many numbers do you actually need for colored light? Assume all tuples (R,G,B) are legitimate colors (they are).

Image Credit: http://en.wikipedia.org/wiki/File:RGB_illumination.jpg

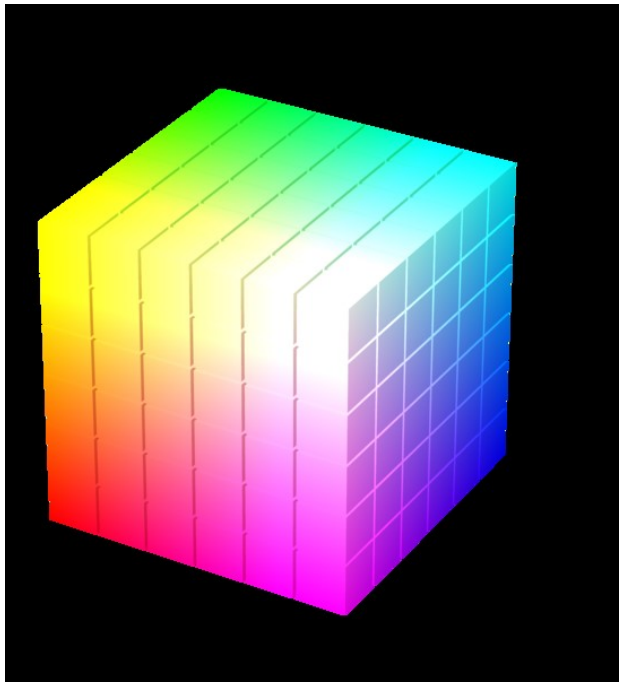
One Option: RGB

<u>Pros</u>

1. Simple
2. Common

<u>Cons</u>

1. Distances don't make sense
2. Correlated



R

G

B

Slide Credit: J. Hays, RGB cube: https://en.wikipedia.org/wiki/RGB_color_model

RGB

Photo credit: J. Hays

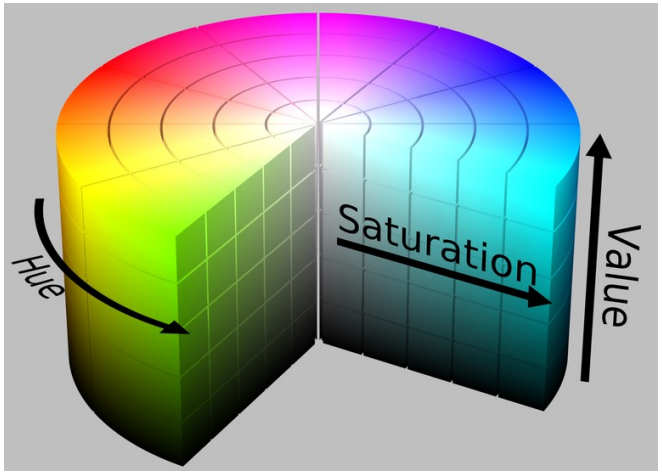
Another Option: HSV

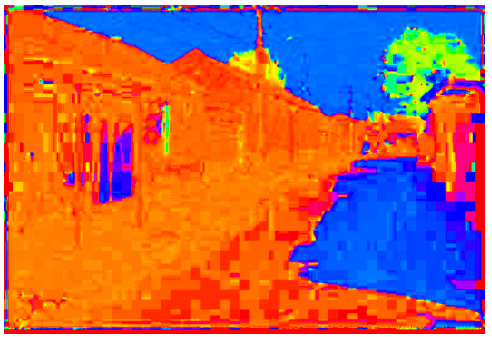
Pros

1. Intuitive for picking colors
2. Sort of common
3. Fast to convert

Cons

1. Not as good as other better spaces





H
(S=1,V=1)

S
(H=1,V=1)

V
(H=1,S=0)

Slide Credit: J. Hays, HSV cylinder: https://en.wikipedia.org/wiki/HSL_and_HSV

HSV

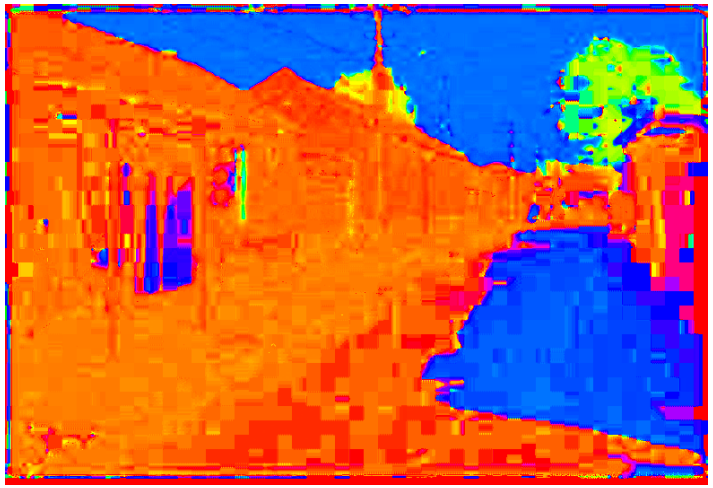


Photo credit: J. Hays

Another Option: YCbCr/YUV

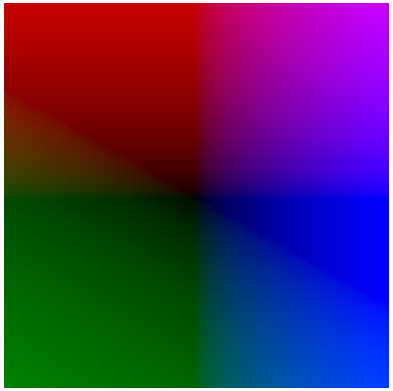
<u>Pros</u>

1. Great for transmission / compression

<u>Cons</u>

1. Not as good as other better smart color spaces

Y = 0



Y = 0.5

Y
(Cb=0.5, Cr=0.5)

Cb
(Y=0.5, Cr=0.5)

Cr
(Y=0.5, Cb=05)

Slide Credit: J. Hays, YUV cube: https://en.wikipedia.org/wiki/YUV

YCbCr

Photo credit: J. Hays

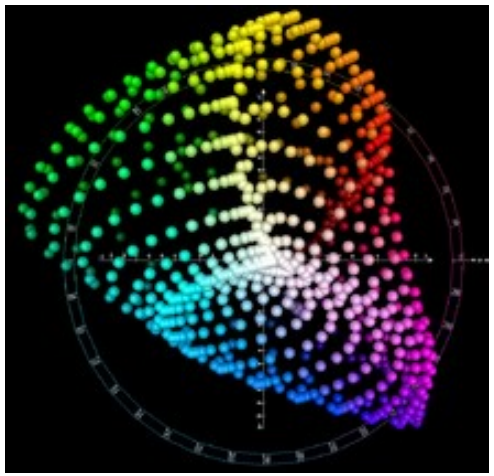
Another Option: Lab

<u>Pros</u>

1. Distances correspond with human judgment
2. Safe

<u>Cons</u>

1. Complex to calculate (don't write it yourself, lots of fp calculations)



L
(a=0,b=0)

a
(L=65,b=0)

b
(L=65,a=0)

Slide Credit: J. Hays, Lab diagram cube: https://en.wikipedia.org/wiki/CIELAB_color_space

Lab

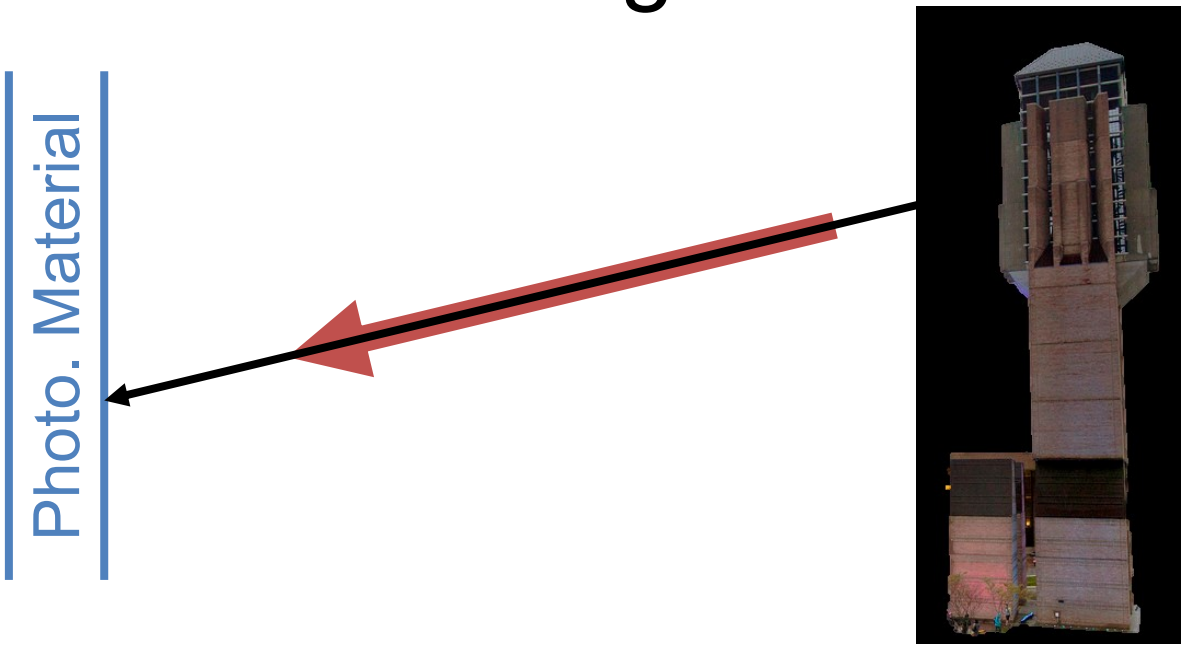
Photo credit: J. Hays

Why Are There So Many?

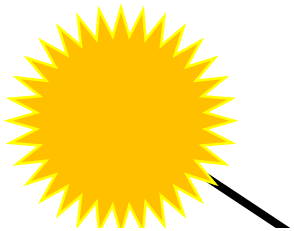
- Each serves different functions
  - RGB: sort of intuitive, standard, everywhere
  - HSV: good for picking, fast to compute
  - YCbCr/YUV: fast to compute
  - Lab: the right(?) thing to do, but “slow” to compute
- Pick based on what you need and don’t sweat it: color really isn’t crucial

So Far

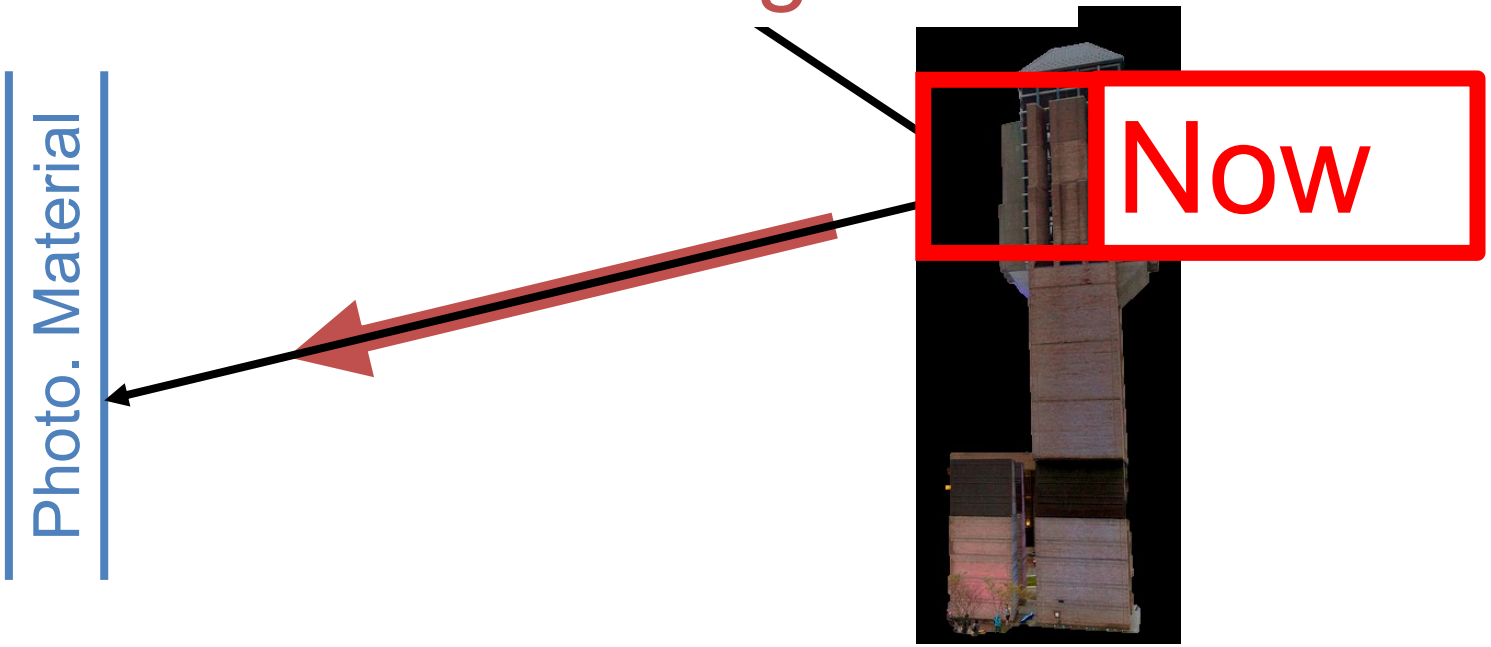
How do we represent light and its storage on film?



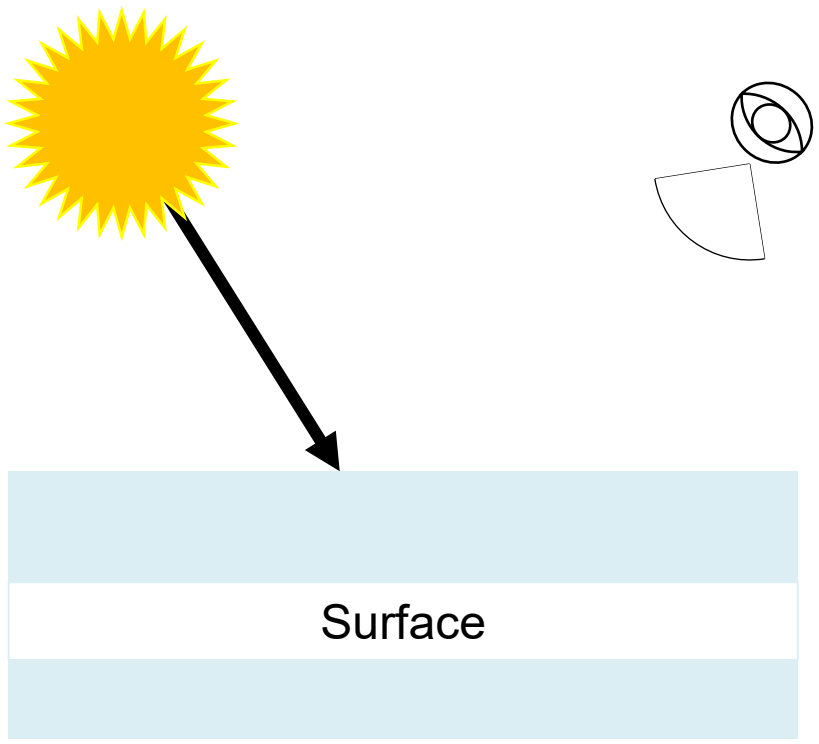
Now



How does the scene cause that light?

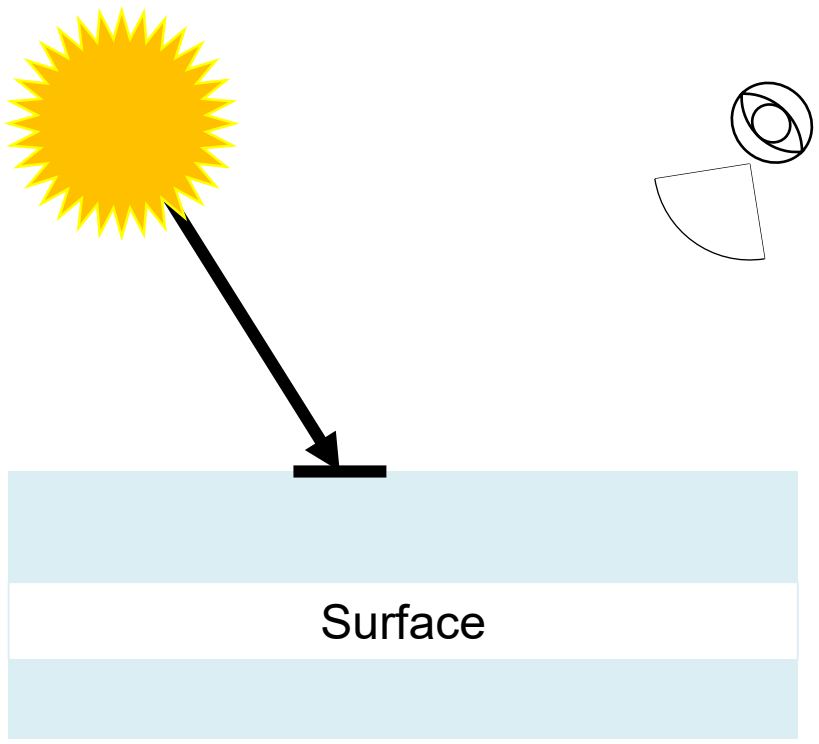


Light and Surfaces



What happens when light hits a surface?

Light and Surfaces

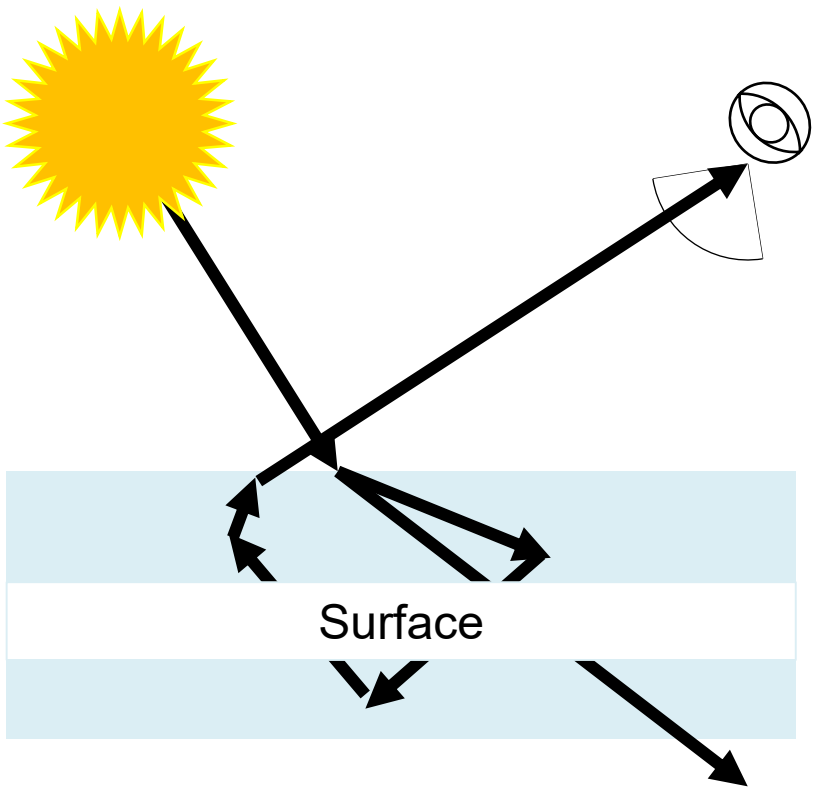


What happens when light hits a surface?

1. Absorbed

It's absorbed and converted into some other form of energy (e.g., a black shirt getting hot in the sun)

Light and Surfaces

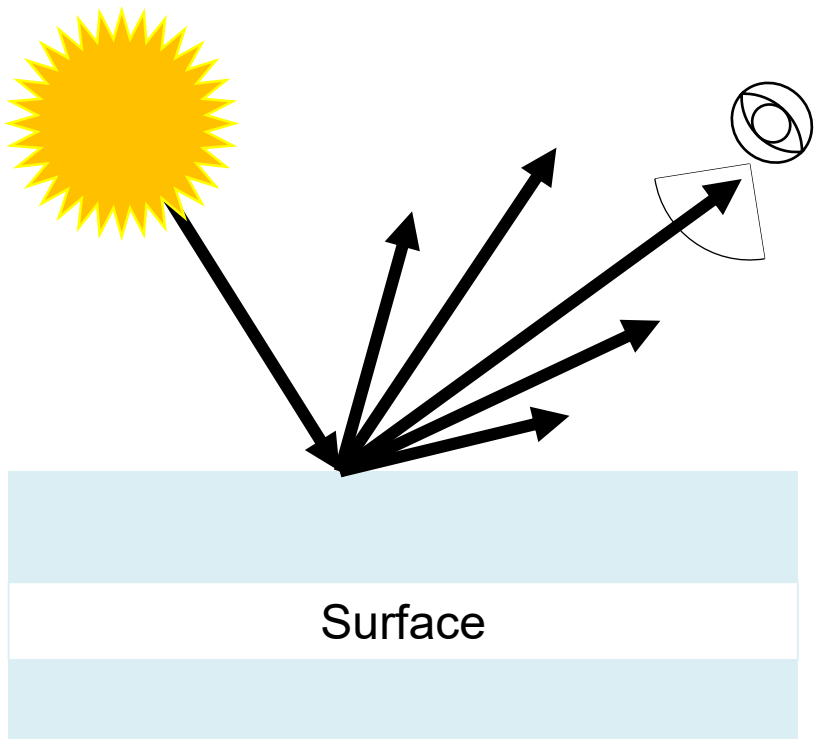


What happens when light hits a surface?

2. Transmitted

Possibly bouncing around before going through or out (e.g. lenses bend and go through, milk bounces around)

Light and Surfaces

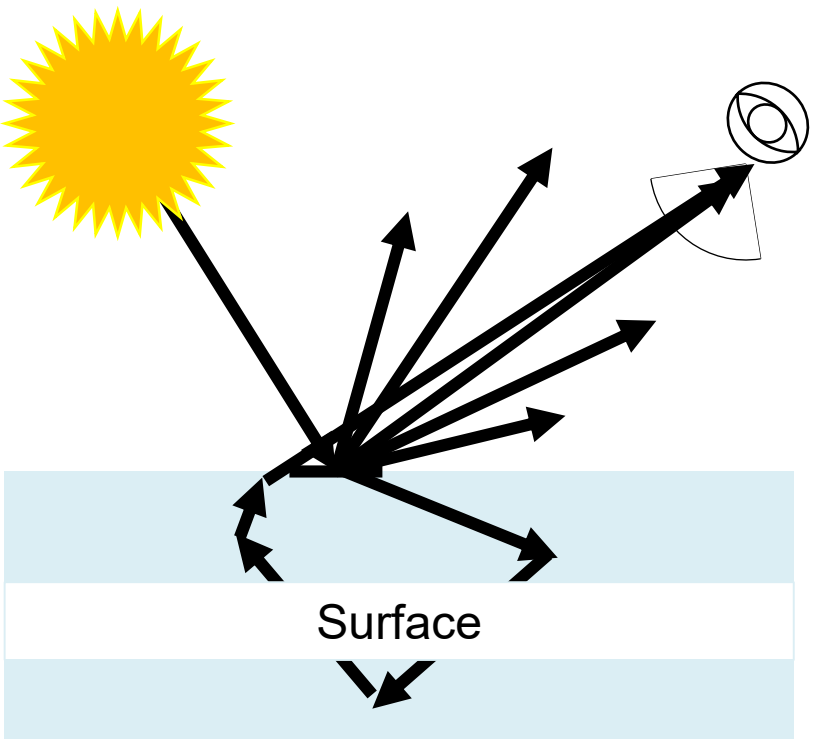


What happens when light hits a surface?

3. Reflected

It's reflected back, in one or more directions with varying amounts (e.g., mirror, or a white surface)

Light and Surfaces

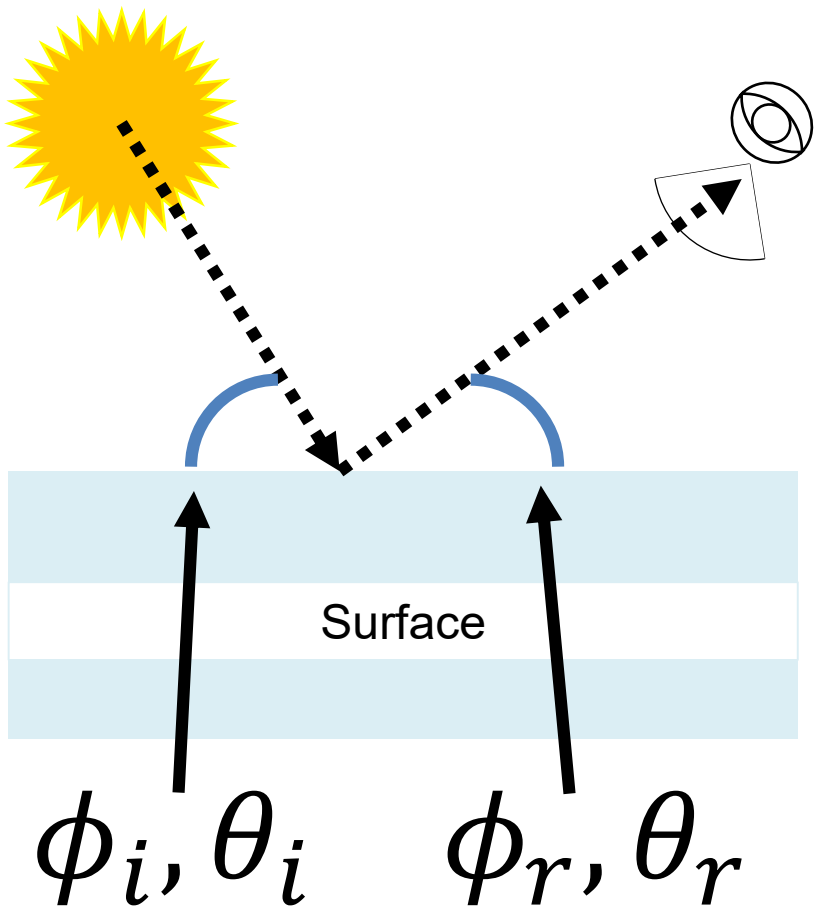


What happens when light hits a surface?

4. Everything

All of the above! Real surfaces often have combinations of all of these options.

Modeling Light and Surfaces



Opaque Reflections

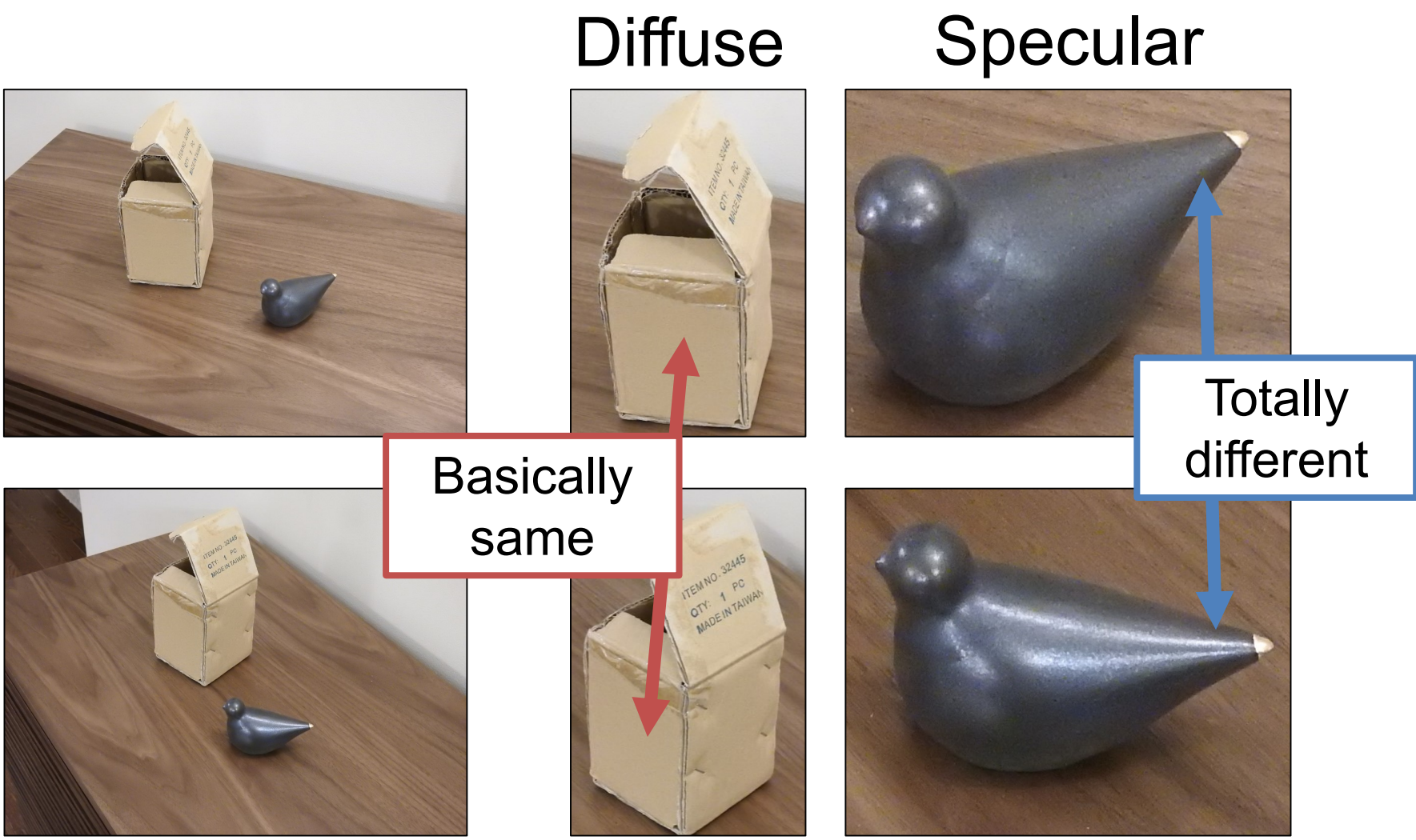
Bi-directional reflectance function: % reflected given ***i****ncident angle* to light ***r****eflected angle* to the viewer.

Note: have not specified form of function.

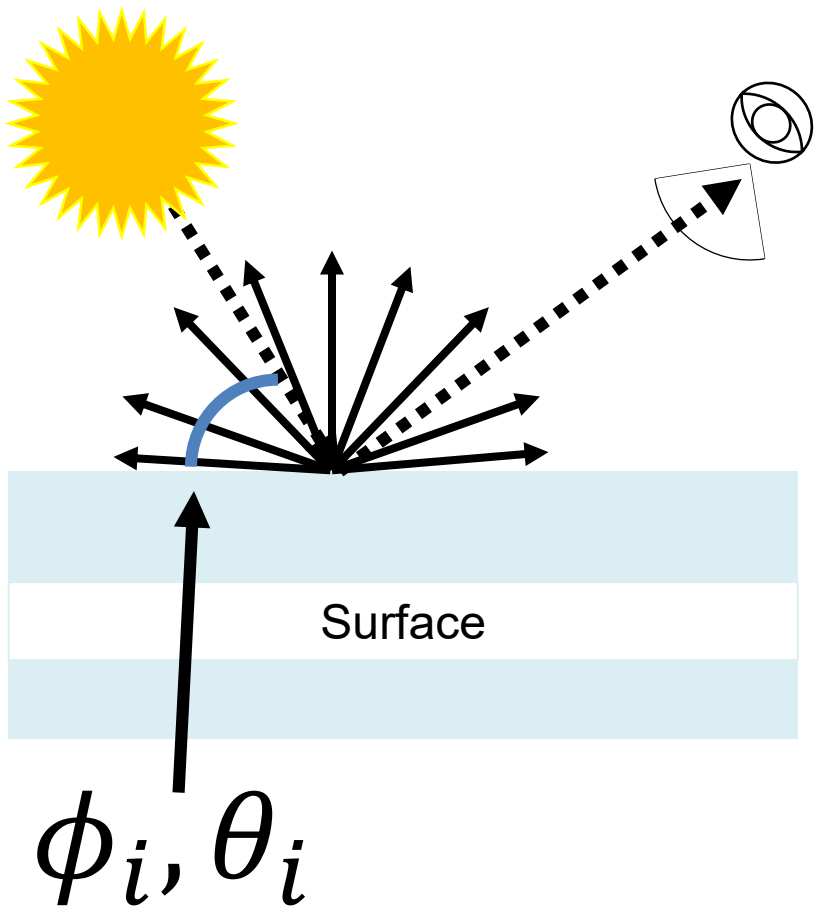
Specular and Diffuse Reflection

Same lighting, as close as possible camera settings, but different **location**

Specular and Diffuse Reflection



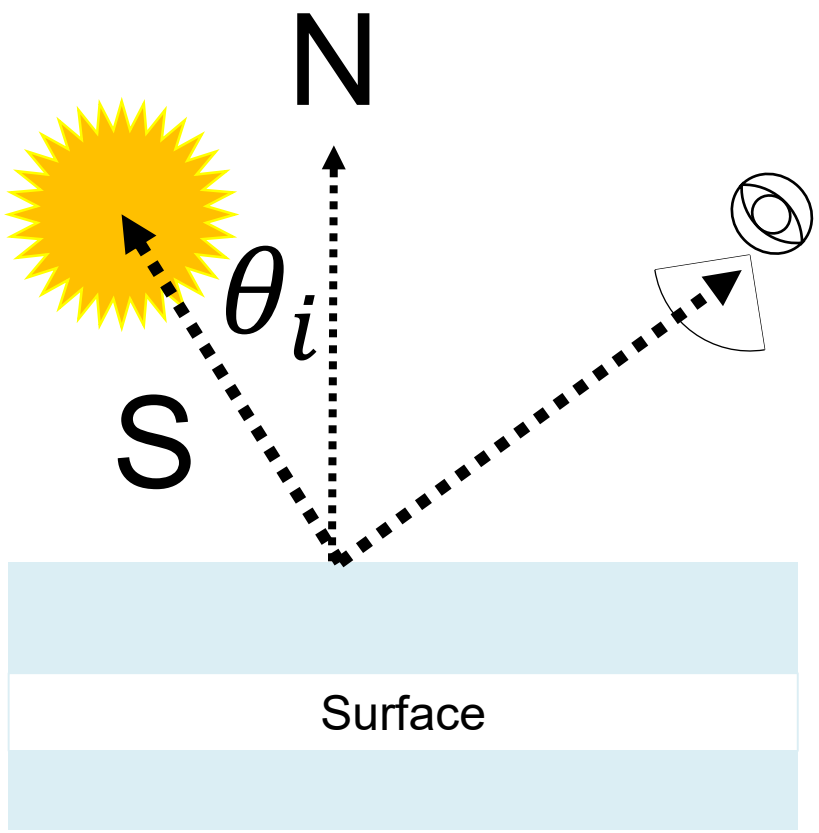
Diffuse Reflection



Lambertian Surface

Light depends **only** on orientation of surface ϕ_i, θ_i to light. Result of random small facets. Looks identical at all views.

Diffuse Reflection



Lambert's Law

N: surface normal

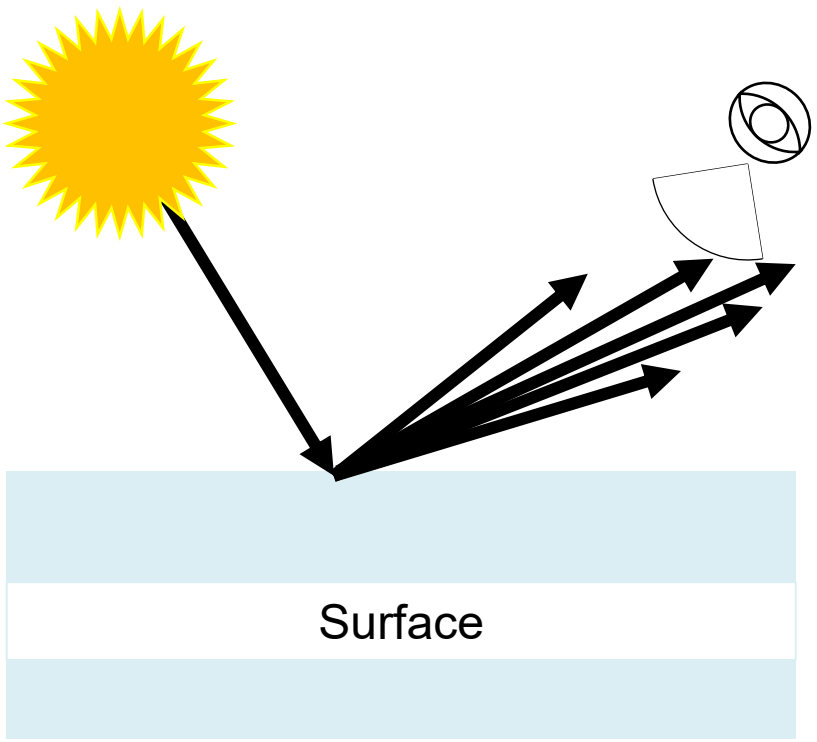
S: source direction **and** strength

ρ: how much is reflected

$$B = \rho \boldsymbol{N} \cdot \boldsymbol{S}$$

$$B = \rho \|\boldsymbol{S}\| \cos(\theta)$$

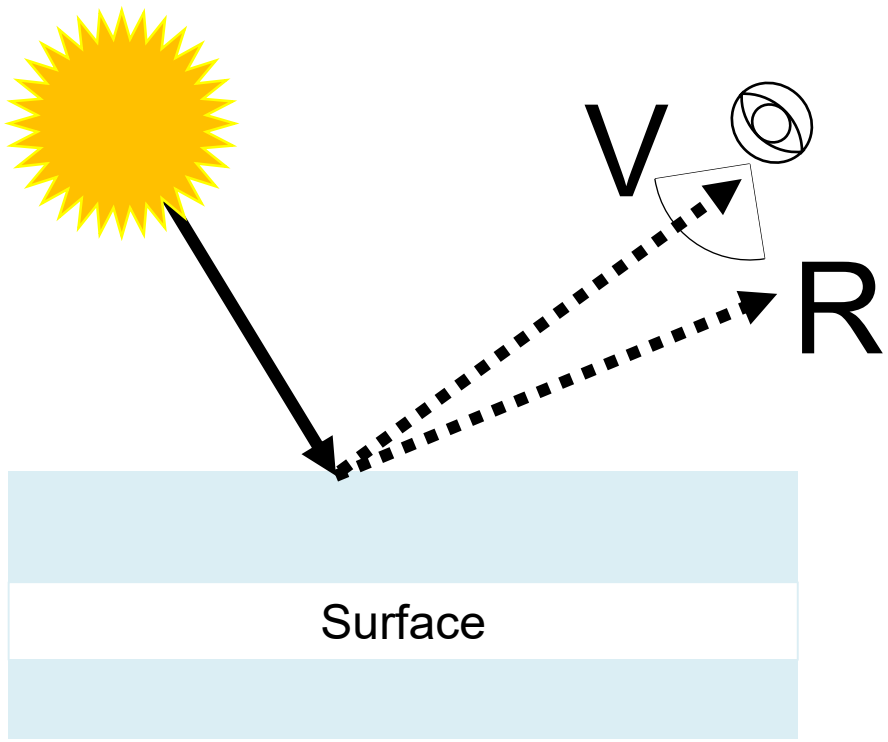
Specular Reflection



Specular Surface

Light reflected like a mirror, but spreads out in a “lobe” around the reflection ray

Specular Reflection



Phong Model

V: angle to viewer

R: reflection ray

α: shininess constant

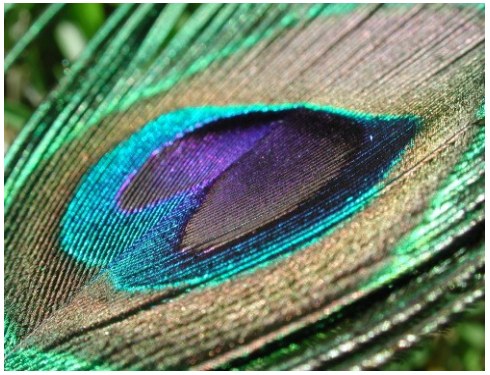
$$B = (V^T R)^{\alpha}$$

BRDFs can be incredibly complicated…










Slide Credit: L. Lazebnik

What Can This Be Used For

Shape from Shading

Lambert's Law: for every pixel i

$$B_i = \rho \boldsymbol{N}_i \cdot \boldsymbol{S}$$

Reflected Light (1 dim)	Surface Orientation (3? dim)	Illumination Global, (3 dim)

Given: illumination and light, recover normals

Potential problems?

Shape From Shading

$$B_i = \rho \boldsymbol{N}_i \cdot \boldsymbol{S}$$

1D, **fixed** actually 2D **unknown** 3D, **fixed**

- System of equations that's underdetermined (N equations, 2N unknowns, N+3 known)
- **Solution**: Add more equations that enforce smoothness or finding a single surface.

Realistic Shape From Shading

$$B_i = \rho \boldsymbol{N}_i \cdot \boldsymbol{S}$$

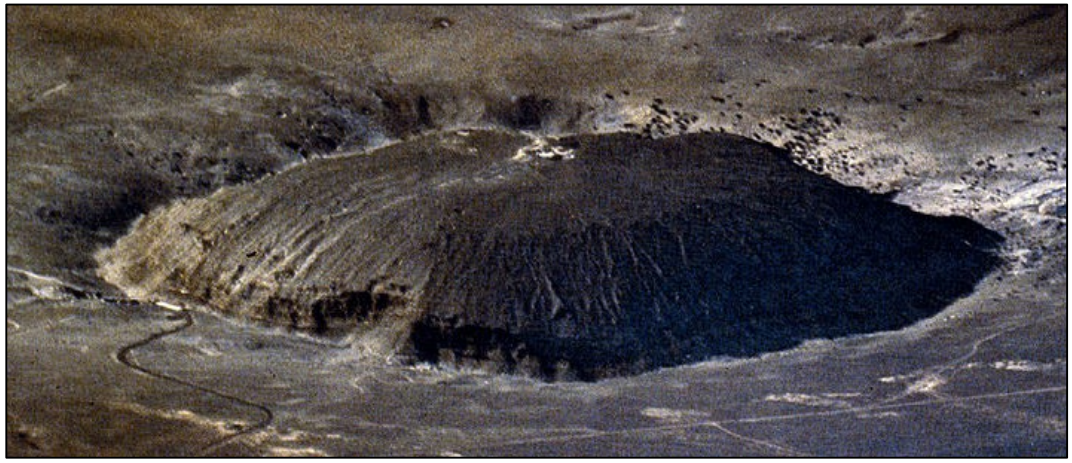
1D, **fixed**

2D **unknown**

3D, **unknown**

- System of equations that's underdetermined (N equations, 2N+3 unknowns)
- **Solution**: need prior beliefs to disambiguate.

Ambiguity

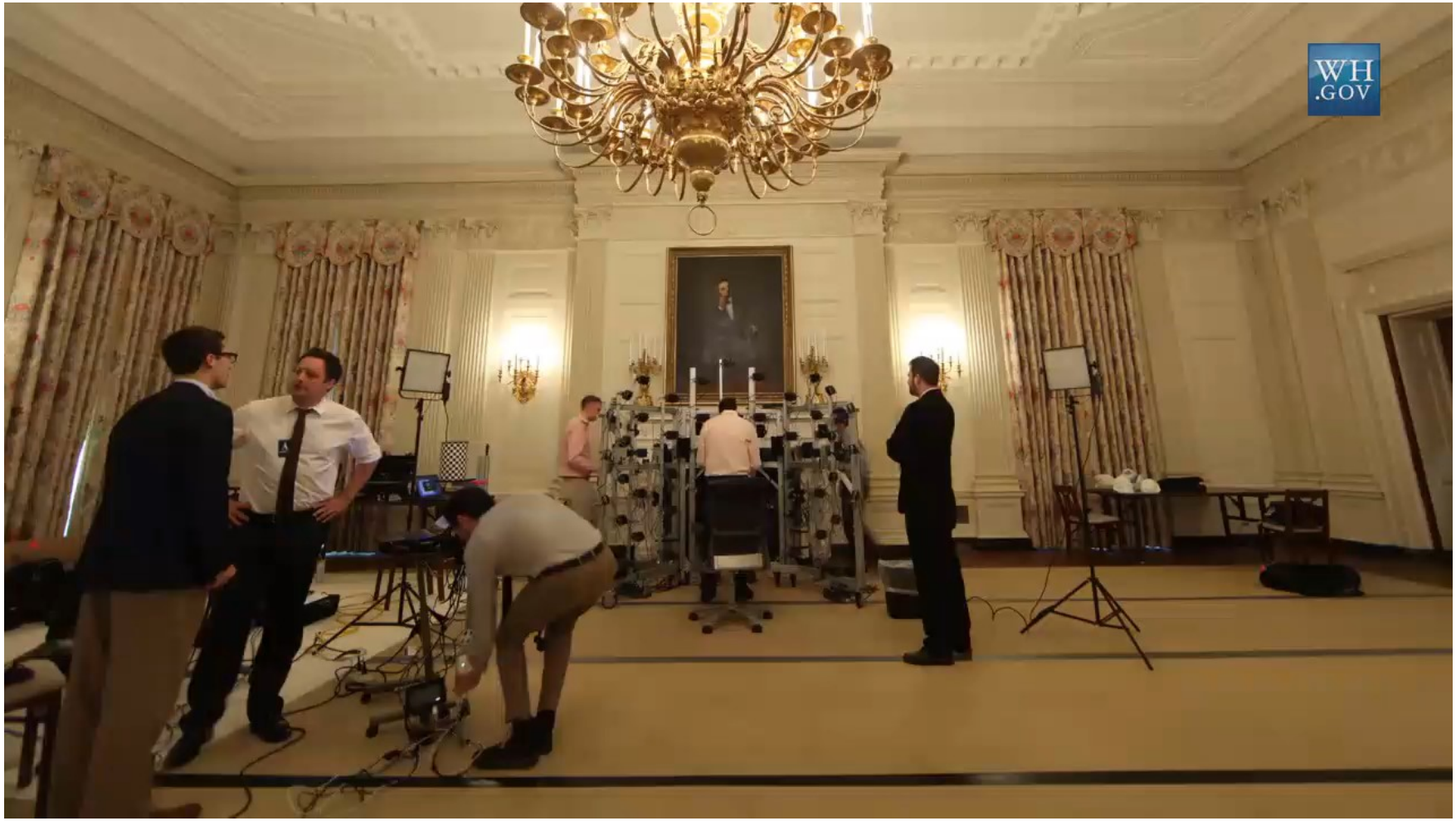


Ambiguity

Humans assume light from above (and the blueness also tells you distance)

Photo Credit: https://en.wikipedia.org/wiki/Meteor_Crater

Shape from Shading in Practive



https://www.youtube.com/watch?v=4GiLAOtjHNo